



# **Solution Approach**

## **in Integration of AI Engine into AGL**

**Japan Technical Jamboree 64**

March 2nd, 2018

NTT DATA MSE Corporation

Hiroto Imamura

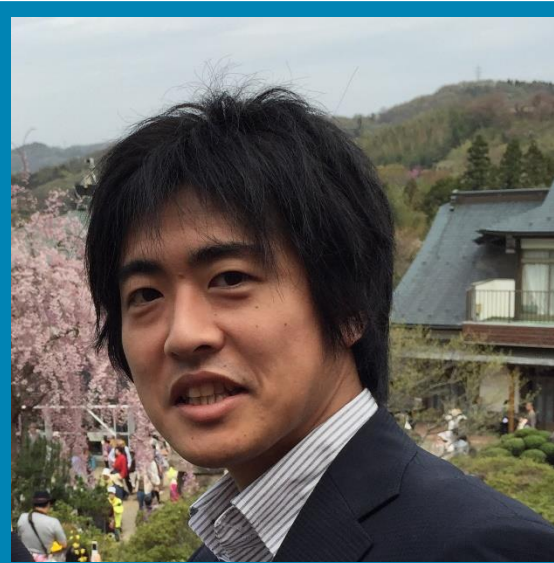
Tomonari Okuno

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# Who are we?

## NTT DATA MSE Corporation



Name	<b>Hiroto Imamura</b>
Position	<b>Manager</b>
Carrier	<ul style="list-style-type: none"><li>- <b>Leader of OSS Collaboration related activities in NTT DATA MSE</b></li><li>- <b>Linux-based embedded devices</b></li><li>- <b>Architecture design / System debugging / Performance optimization / Security</b></li></ul>



Name	<b>Tomonari Okuno</b>
Position	<b>Deputy Manager</b>
Carrier	<ul style="list-style-type: none"><li>- <b>Lead Architect of R&amp;D projects in NTT DATA MSE</b></li><li>- <b>Linux-based embedded devices</b></li><li>- <b>Performance optimization / OSS Licenses</b></li></ul>

# AUTOMOTIVE GRADE LINUX

## About

Automotive Grade Linux (AGL) is a collaborative open source project that is bringing together automakers, suppliers and technology companies to build a Linux-based, open software platform for automotive applications that can serve as the de facto industry standard. Adopting a shared platform across the industry reduces fragmentation and allows automakers and suppliers to reuse the same code base, leading to rapid innovation and faster time-to-market for new products.

As a “code first” organization, AGL’s goals are to:

- Build a single platform for the entire industry
- Develop 70-80% of the starting point for a production project
- Reduce fragmentation by combining the best of open source
- Develop an ecosystem of developers, suppliers, expertise all using a single platform

Although initially focused on infotainment, AGL is the only organization planning to address all software in the vehicle: infotainment, instrument cluster, heads-up-display (HUD), telematics/ connected car, advanced driver assistance systems (ADAS), functional safety and autonomous driving.

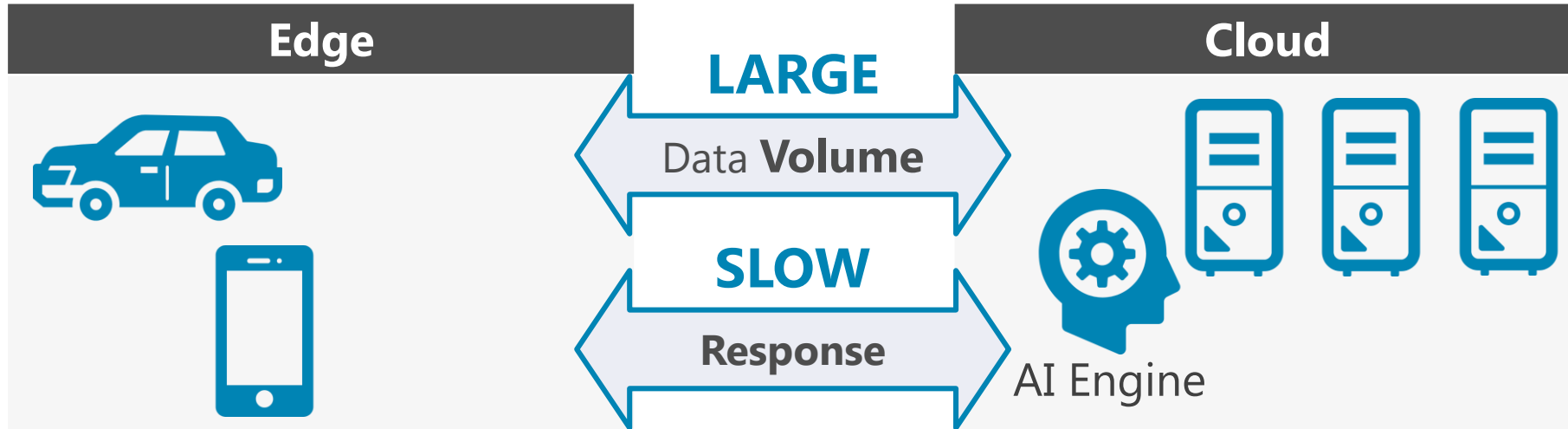
Automotive Grade Linux is a Project at [The Linux Foundation](https://www.linuxfoundation.org/).

Ref.

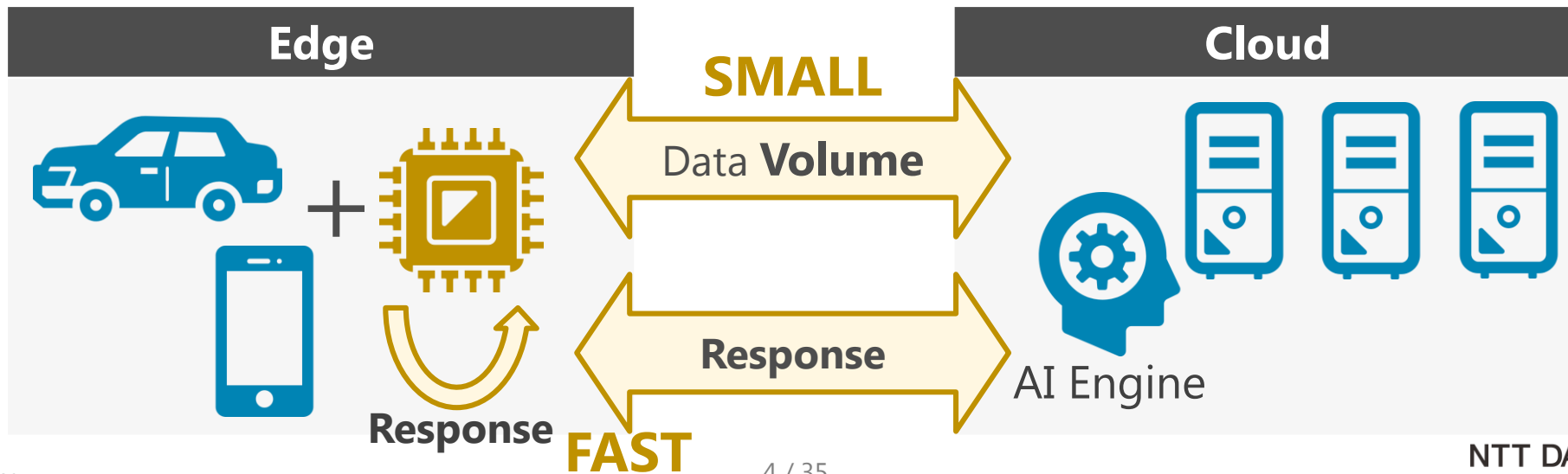
- <https://www.automotivelinux.org/about>
- <https://www.automotivelinux.org>

# Why Edge AI ?

So far



Near future



# Edge AI is spotlighted

More and more Edge AI solutions have been announced

Apr. 2017	<b>e-AI</b> by Renesas Electronics
Jun. 2017	<b>Neural Network Libraries</b> by Sony
Jun. 2017	<b>Embedded Learning Library</b> by Microsoft
Jul. 2017	<b>TensorFlow Lite</b> by Google
Jul. 2017	<b>revision</b> by Xilinx
Sep. 2017	<b>Kirin 970</b> by HUAWEI
Jan. 2018	<b>Deep Learning Accelerator Card</b> by PFU

## Our motivation

1. OSS / Rich development environment **Easy to try**
2. Few people addressed **Good chance to appeal our technical capabilities**
3. AGL Member **Make collaboration**

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# Collaboration



supported by Sony members

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# Feedback at CES 2018



Valuable opinions, Valuable discussions

# Technical Explanation



## ■ Our team have started AI related activities from October 2017

- There are AI related news almost everyday
- Our interest:
  - ✓ How is the performance of AI on edge devices?
  - ✓ What do we need to learn in order to realize AI on edge devices?
- First Step: Let's use an AI engine on edge devices
- Implemented a demo system "Handwritten Digit Recognition App"

## ■ In this presentation

- Overview of Machine Learning on edge devices
- How we implemented the "Handwritten Digit Recognition App" on AGL

# What is AI?

## What are the relations of these terms?

### Artificial Intelligence

Computer programs to simulate functions of human's brain such as "cognition", "judgement", etc

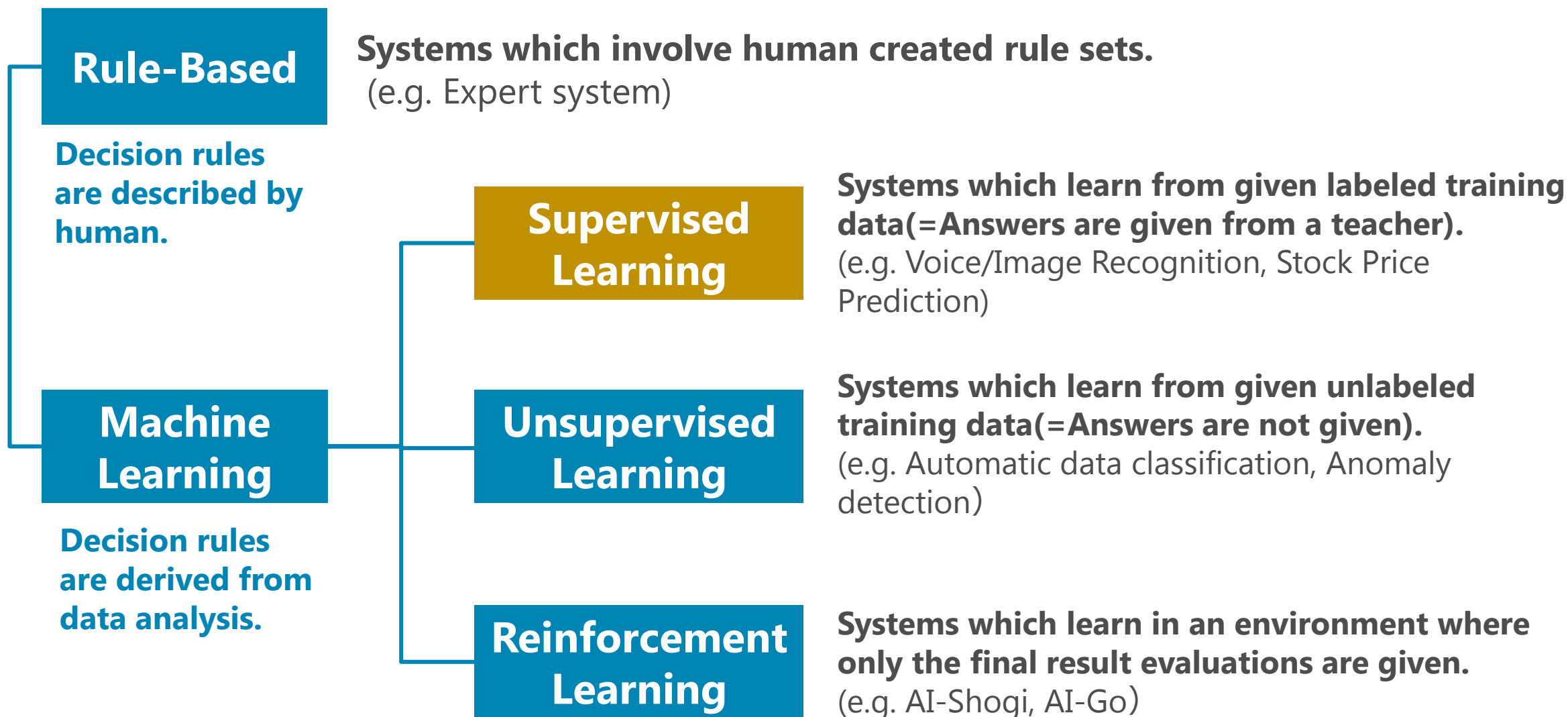
### Machine Learning

Technology to let computers learn rules and knowledge from vast amount of various data such as values, texts, pictures and voice.

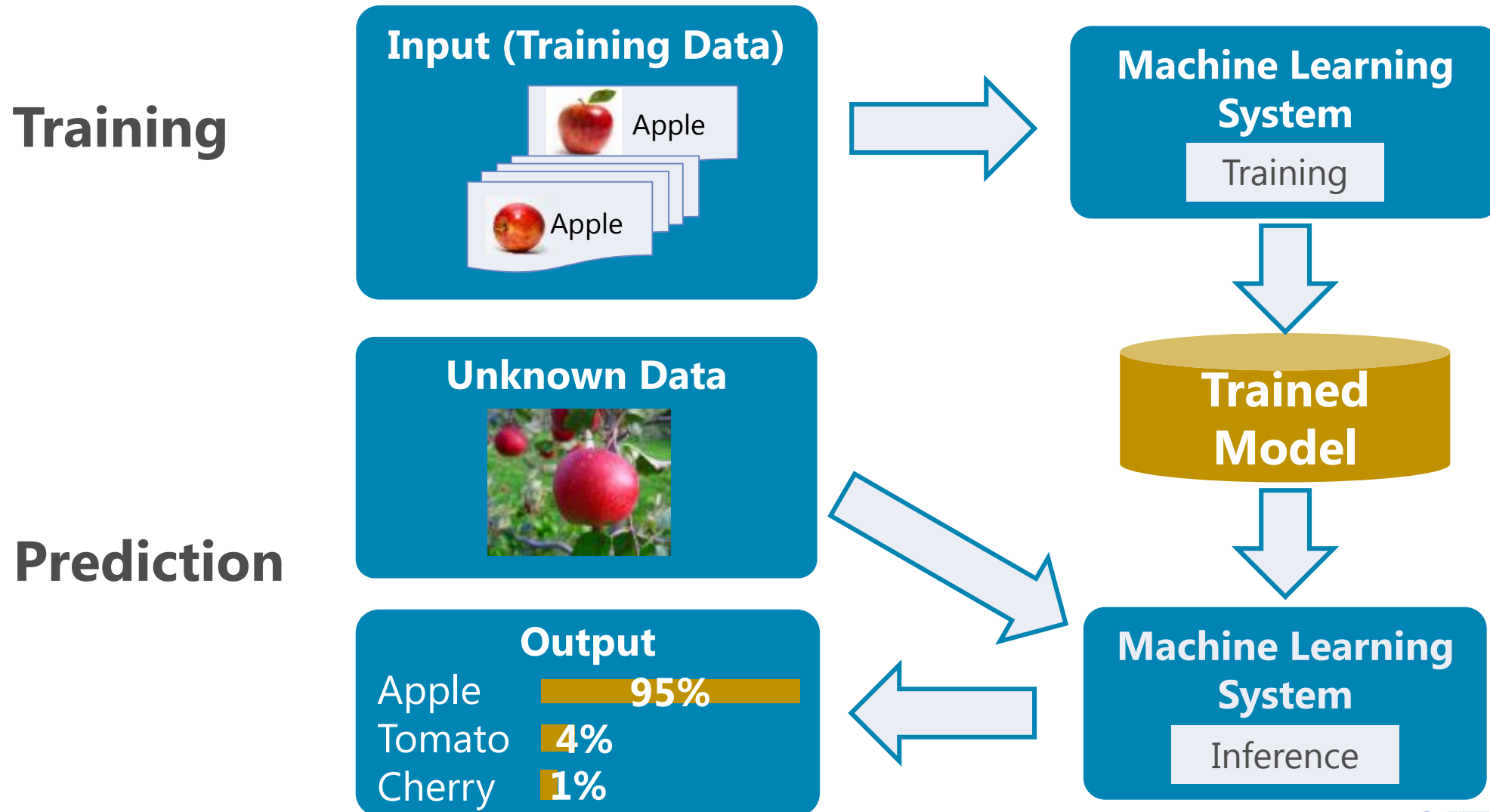
### Deep Learning

One of the method to perform machine learning by using neural network to realize high level of abstraction by extracting information from multiple layers one by one.

# Types of AI



# Machine Learning (Supervised learning)



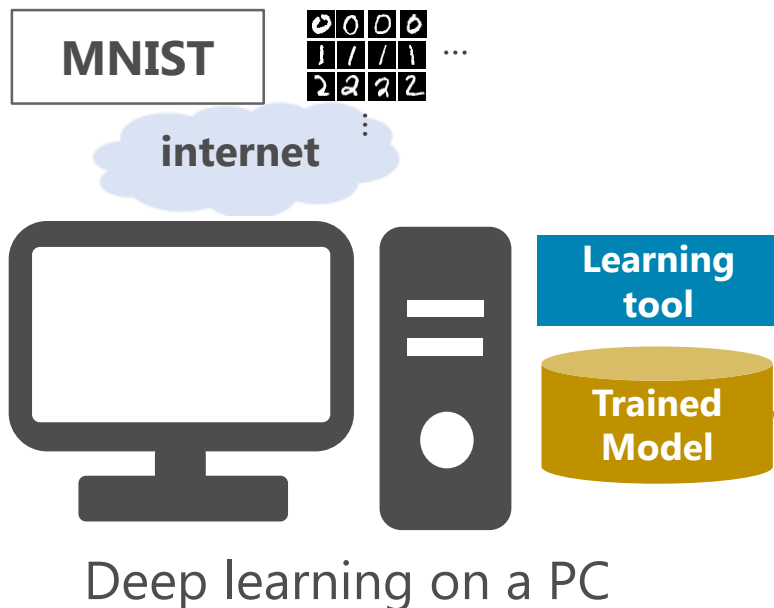
# Demo System: Handwritten Digit Recognition App



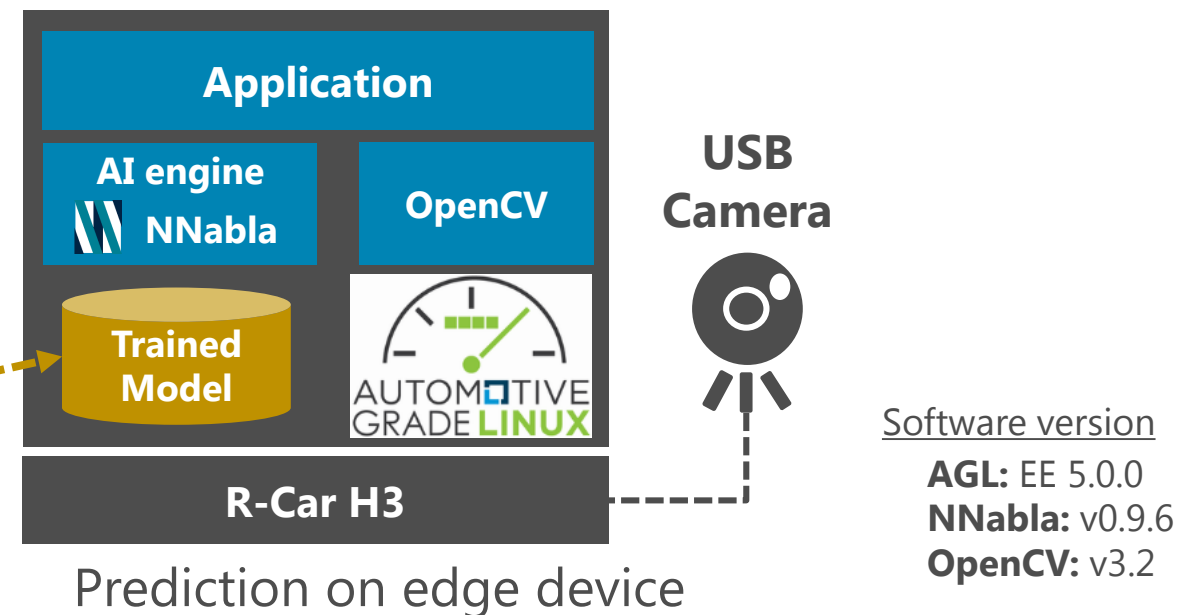
# System Overview

- Recognizes handwritten digits from an image captured with an USB Camera
- Deployed a pre-trained model to an edge device

## Training



## Prediction



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# Application GUI

**Recognized Area**  
(100 x 100 pixels)

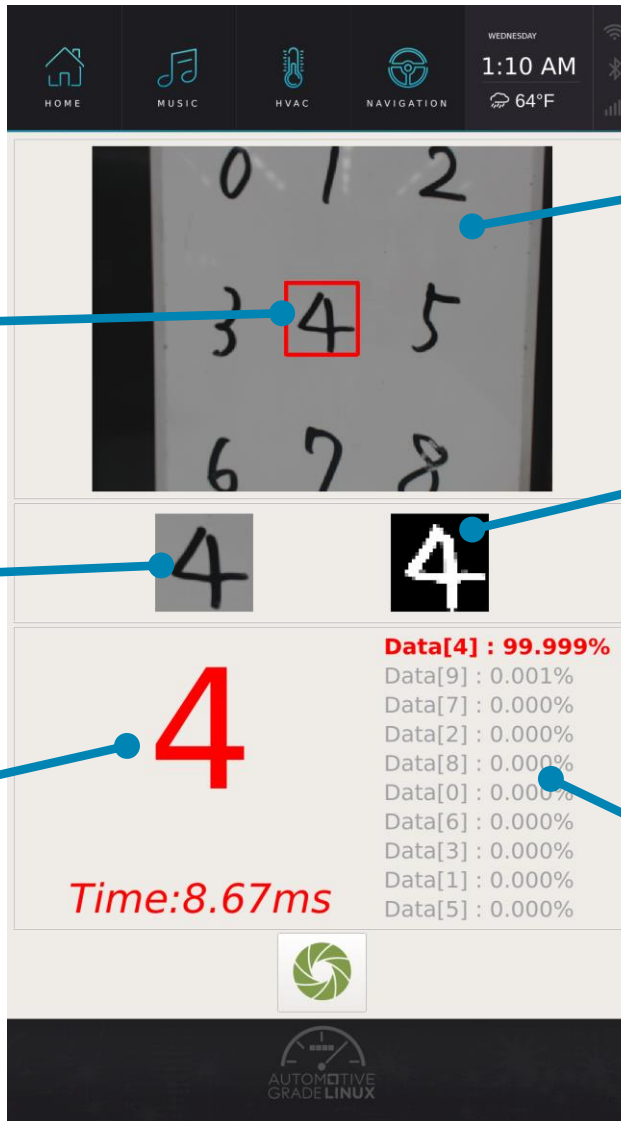
**Cropped image**  
(100 x 100 pixels)

**Recognized digit**

**USB Camera image**

**Input image  
to AI Engine**  
(28 x 28 pixels)

**Prediction result  
from AI Engine**  
(Probability of each digit)



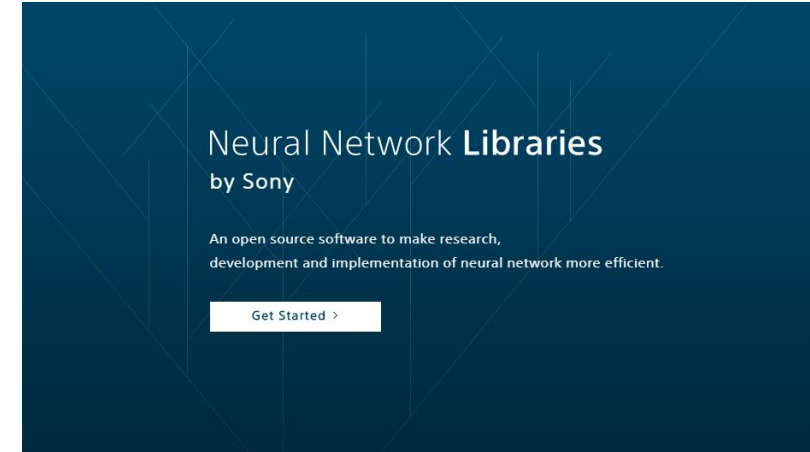
# Neural Network Libraries (NNabla)

## Neural Network Libraries

<https://nnabla.org>



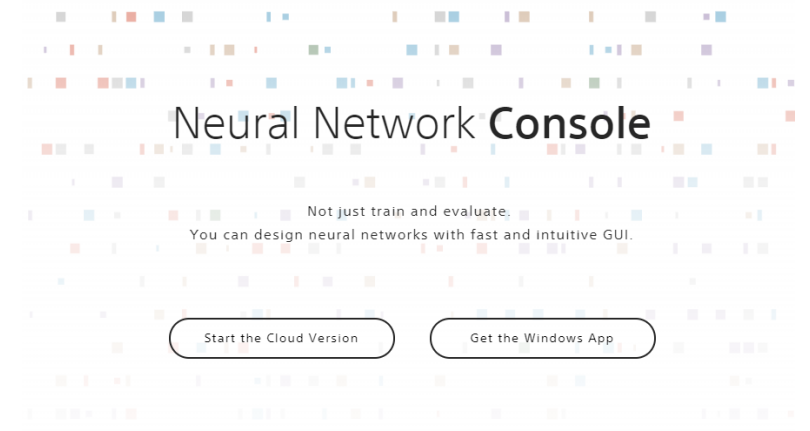
- Deep learning framework.
- Intended to be used for research, development and production.
- Aim to have it running everywhere.  
Deployable to embedded devices.
- Apache License 2.0



## Neural Network Console

<https://dl.sony.com>

- GUI tool for designing neural networks intuitively.
- Many useful functions to support research and development.
- Trained model can be embedded by using Neural Network Libraries.



# Details of the Implementation

**Step1:** Deep Learning on PC

**Step2:** Building / Installation

**Only for DD**

**Step3:** Enabling USB webcam

**Only for DD**

**Step4:** Installation of the OpenCV

**Step5:** App Implementation

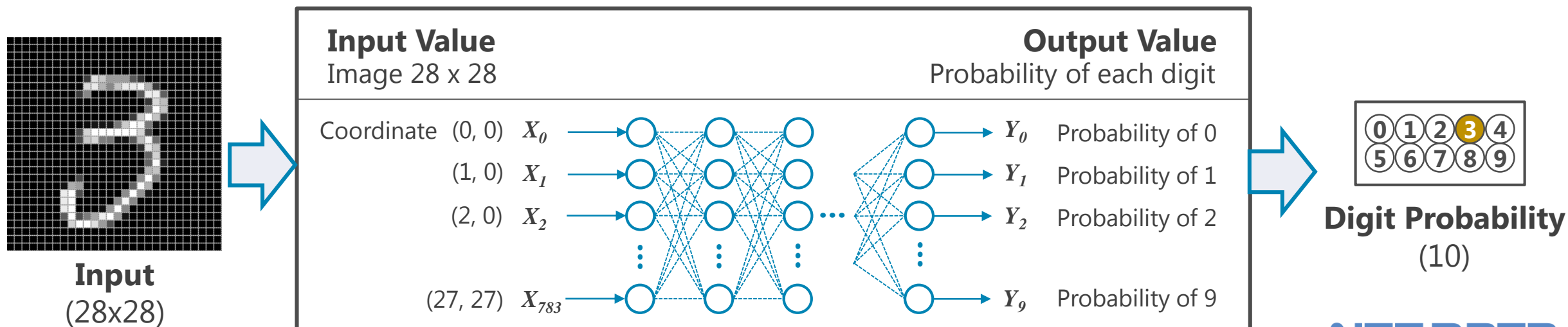
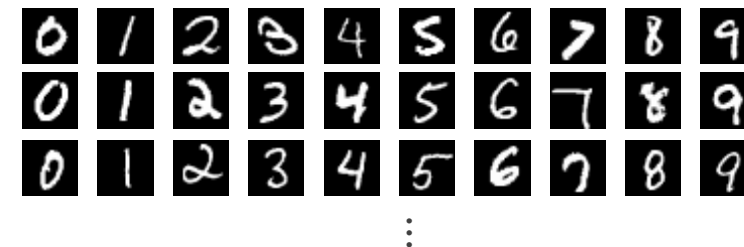
Step3 and 4 are not needed if AGL version is 5.0(EF)

# Step1: Deep Learning on PC

Used one of the examples available at the github

→ **mnist-collection/classification.py**

- Downloads and uses MNIST Dataset as training data (60,000 samples of handwritten digits image and label)
- Uses Convolutional Neural Network
  - Input: 28 x 28 pixels grayscale image
  - Output: Prediction of 10-way classification



# Step1: Deep Learning on PC

Used one of the examples available at the github  
→ [mnist-collection/classification.py](https://github.com/sony/nabla-examples/tree/master/mnist-collection/classification.py)

## Install the NNabla on a PC

```
$ pip install nnabla
```

## Obtain the “mnist-collection”

```
$ cd ~/work/sony/  
$ git clone https://github.com/sony/nabla-examples
```

Obtain examples

## Start training

```
$ cd nnabla-examples/mnist-collection  
$ python classification.py  
2018-01-30 19:42:37,932 [nnabla][INFO]: Initializing CPU extension...  
.....  
2018-01-30 19:42:38,437 [nnabla][INFO]: Using DataIterator  
2018-01-30 19:42:39,343 [nnabla][INFO]: iter=9 {Training loss}=2.30425691605  
2018-01-30 19:42:39,343 [nnabla][INFO]: iter=9 {Training error}=0.8375  
.....  
2018-01-30 19:53:18,056 [nnabla][INFO]: Parameter save (.h5): tmp.monitor/lenet_params_010000.h5
```

Start

Complete



# Step2: Building / Installation

## Build NNabla for R-Car H3(ARMv8(64bit))

### Install cross SDK on a PC (AGL R-Car ARMv8 toolchain)

```
$ wget https://download.automotivelinux.org/AGL/release/eel/5.0.0/m3ulcb-nogfx/deploy/sdk/poky-agl-glibc-x86_64-agl-image-ivi-crosssdk-aarch64-toolchain-5.0.0.sh
$ chmod a+x poky-agl-glibc-x86_64-agl-image-ivi-crosssdk-aarch64-toolchain-5.0.0.sh
$ ./poky-agl-glibc-x86_64-agl-image-ivi-crosssdk-aarch64-toolchain-5.0.0.sh
```

See also:

[http://docs.automotivelinux.org/docs/getting\\_started/en/dev/reference/source-code.html](http://docs.automotivelinux.org/docs/getting_started/en/dev/reference/source-code.html)

### Build NNabla for R-Car

```
$ git clone https://github.com/sony/nnabla
$ source /opt/poky-agl/5.0.0/environment-setup-aarch64-agl-linux
$ mkdir -p nnabla/build && cd nnabla/build
$ cmake .. -DBUILD_CPP_UTILS=ON -DBUILD_PYTHON_PACKAGE=OFF
$ make
$ ls -l lib/
-rwxrwxr-x 1 nttmse nttmse 191170296 12月 20 06:21 libnnabla.so
-rwxrwxr-x 1 nttmse nttmse 25392344 12月 20 06:21 libnnabla_utils.so
```

Obtain source code

Setup environment

Build

See also:

<https://nnabla.readthedocs.io/en/latest/cpp/installation.html>

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## Step2: Building / Installation

Install the built shared libraries of the NNabla and the pre-trained model to the target filesystem.

### Shared libraries of the NNabla

```
$ export SDCARD=/tmp/agl
$ sudo mount /dev/sdc1 $SDCARD
$ sudo cp libnnabla.so $SDCARD/usr/lib/
$ sudo cp libnnabla_utils.so $SDCARD/usr/lib/
```

Copy

### Pre-trained model

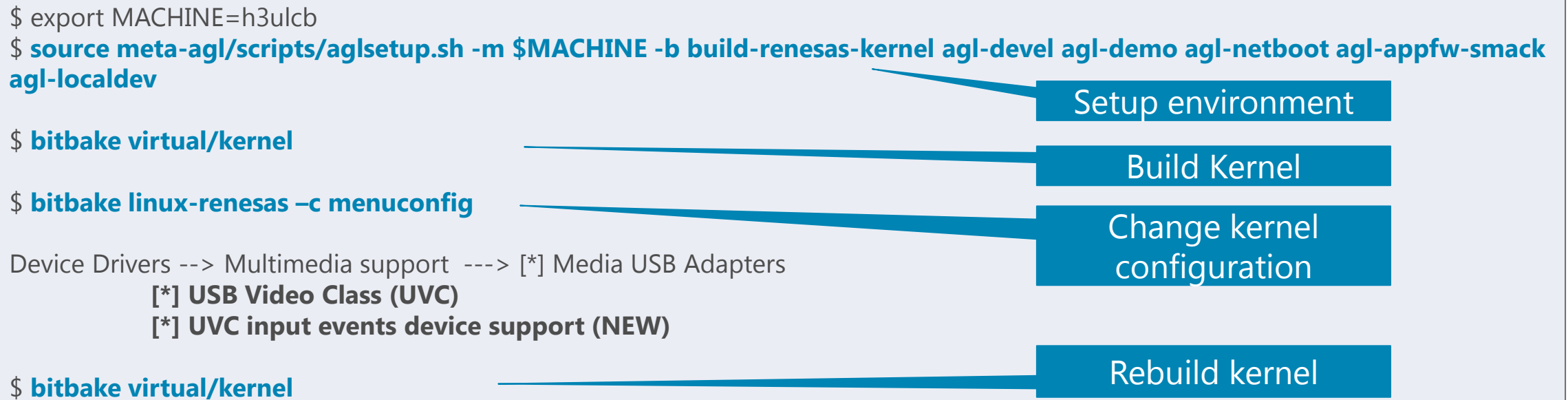
```
$ cd ~/work/sony/nnabla/examples/cpp/mnist_runtime
$ NNABLA_EXAMPLES_ROOT=~/work/sony/nnabla-examples python save_nnp_classification.py
$ ls -l lenet_010000.nnp
-rw-rw-r-- 1 nttmse nttmse 86920 1月 29 19:16 lenet_010000.nnp
$ sudo cp lenet_010000.nnp $SDCARD/home/data/
$ sync
$ sudo umount $SDCARD
```

Convert to NNabla  
file format (NNP)

Copy

## Enable the USB Video Class in order to use USB webcam on R-Car H3.

### Enable the UVC (USB Video class) of kernel config



### Update kernel image

```
$ sudo cp tmp/deploy/images/m3ulcb/Image--4.9.0+git0+098ccf1c9b-r1-m3ulcb-20171116044641.bin SDCARD/boot/Image-4.9.0-yocto-standard
```

## Install the OpenCV to handle camera images

- OpenCV = Open Source Computer Vision Library
- Used for obtaining image from the webcam and pre-process the acquired images before passing them to the NNabla.

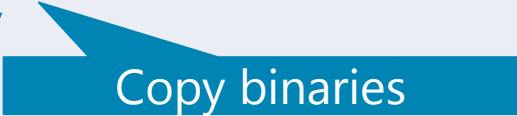
### Build the OpenCV

```
$ export MACHINE=h3ulcb  
$ source meta-agl/scripts/aglsetup.sh -m $MACHINE -b build-opencv agl-devel agl-demo agl-netboot agl-appfw-smack agl-localde  
$ bitbake opencv
```

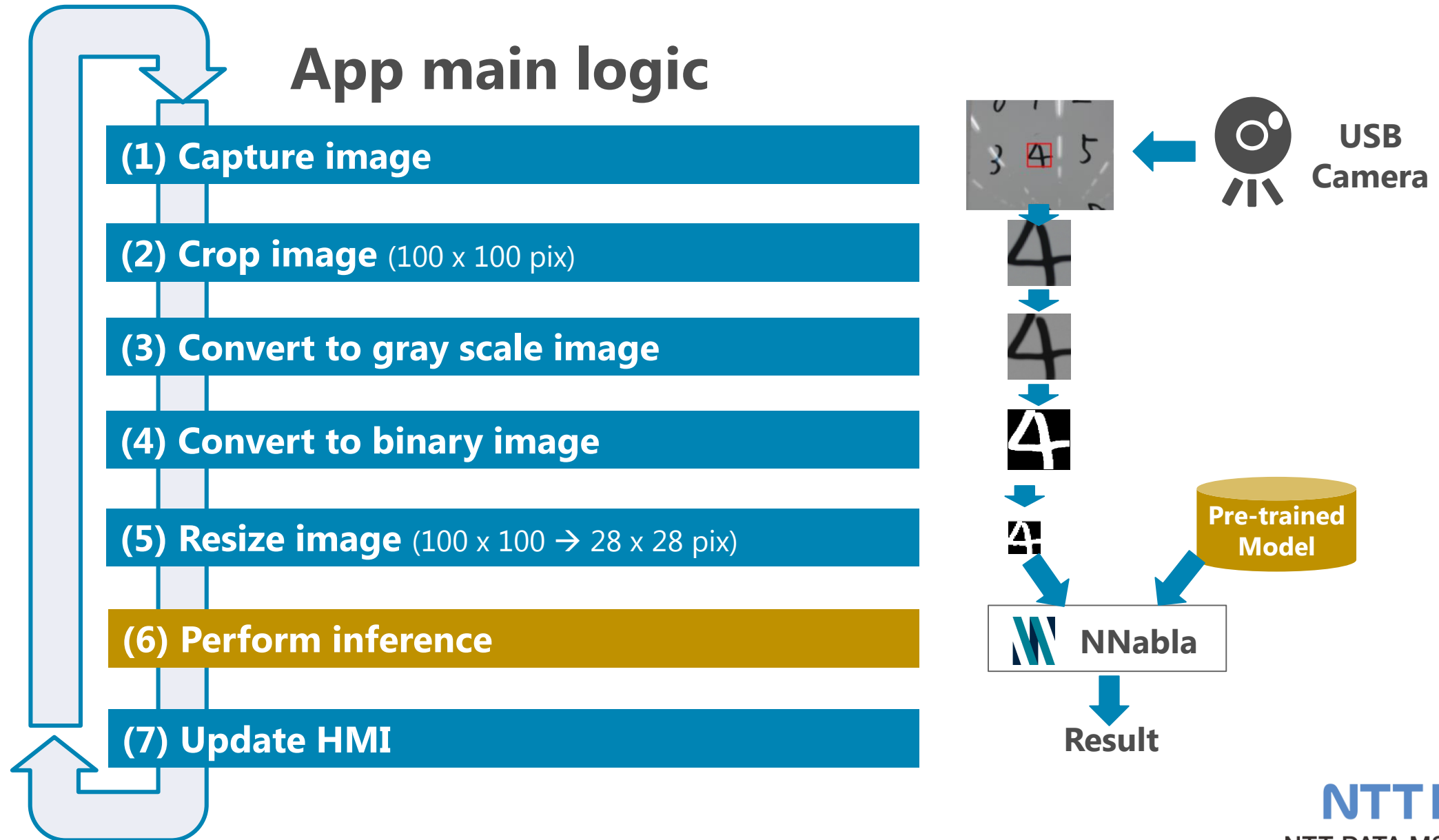


### Install built binaries to target filesystem

```
$ sudo cp -a ./tmp/work/aarch64-agl-linux/opencv/3.2+gitAUTOINC+70bbf17b13-r0/image/* $SDCARD/  
$ sudo cp -a ./tmp/work/aarch64-agl-linux/v4l-utils/1.12.3-r0/image/* $SDCARD/  
$ sudo cp -a ./tmp/work/aarch64-agl-linux/libwebp/0.6.0-r0/image/* $SDCARD/  
$ sync; sudo umount $SDCARD
```



# Step5: App Implementation



# Step5: App Implementation(Source Code)

```
// (0) Open Device
cv::VideoCapture cap("/dev/video22", cv::CAP_V4L2); // videoN: Set according to the environment. This is R-Car H3 with AGL5.0.0.

// (1) Capture image
cv::Mat frame;
cap >> frame;

// Pre-process image
// (2) Crop video image [100x100pix]
cv::Rect rect(GET_VIEW_SIZE_LEFT, GET_VIEW_SIZE_TOP, GET_VIEW_SIZE_WIDTH, GET_VIEW_SIZE_HEIGHT);
cv::Mat rectImg(frame, rect);

// (3) Convert to gray scale image
cv::Mat grayImg;
cv::cvtColor(rectImg, grayImg, CV_RGB2GRAY);

// (4) Convert to binary image ( Invert | Threshold = 127)
cv::Mat binImg;
cv::threshold(grayImg, binImg, 127, 255, cv::THRESH_BINARY_INV);

// (5) Resize image [100x100pix -> 28x28pix]
cv::Mat resizeImg;
cv::resize(binImg, resizeImg, cv::Size(), PGM_WIDTH/grayImg.cols, PGM_HEIGHT/grayImg.rows);

// Add pgm header
pgmformat((char *)"cap.pgm",resizeImg);

// (6) Perform inference
int prediction = 0;
float score[10] = {};
double elapsed = 0;
const std::string nnp_filepath[] = "/home/data/lenet_010000.nnp";
const std::string pgmImageName[] = "cap.pgm";
prediction = EdgeAiNNblaMnistRuntime(pgmImageName, nnp_filepath, score, &elapsed); // nnabla wrapper API
```

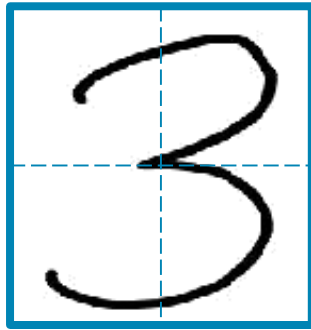


# Experiment Result

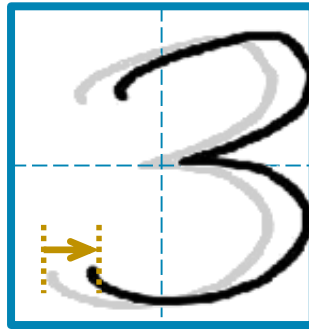
# Recognition Result (First Try)

Not possible to recognize digits for the following patterns

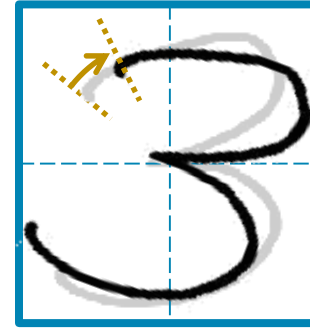
Recognized



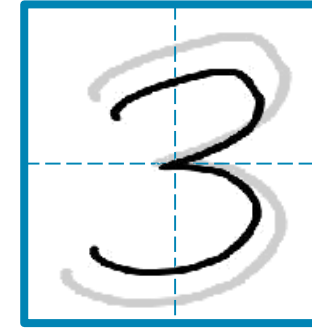
Not Recognized



Shifted



Rotated




Scaled

# Recognition Result (After improvement)

## ■ Used **Data Augmentation** to improve pre-trained model

- Randomly alters the MNIST digit images when perform training.
  - scaling, rotation, aspect ratio, distorting, brightness, contrast, add noise
- Updated the pre-trained model on the device.
- Recognition rate has been improved.
  - Tested with 1,000 cases. (Handwritten digits by our colleague)
  - Counted as "Recognized" if the probability is larger than 50%

Before		After
<b>86.0%</b>		<b>94.8%</b>
(860/1,000)		(948/1,000)

## ■ Improvement of predictions are possible by updating pre-trained models

- Delivering up-to-date pre-trained models to devices can provide more accurate prediction results for users.

## Processing Time



\*Measured on R-Car H3

## Resource Usage

#	Item	Value
1	CPU usage rate	55 %
2	Memory [RSS] usage	80 MB
	Size of the Pre-trained Model	85 KB

\*Measured on R-Car H3

# Conclusion

## ■ Explained how we implemented the Handwritten Digit Recognition App

- Performed deep learning on a PC
- Deployed pre-trained model to R-Car H3 board
- Variation of training data set affects inference results
- Updating of pre-trained model can improve inference results

## ■ It wasn't difficult than expected to take the first step toward using AI engine on AGL

- Implemented the app in 3 weeks
- Please try it!



## ■ Next Step

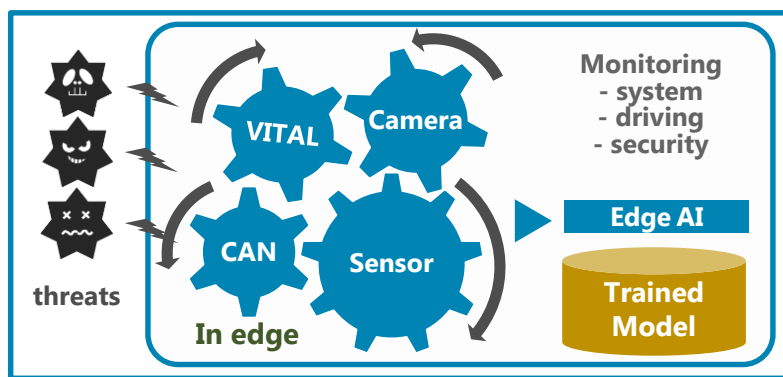
### 1. Use other types of data for deep learning

- Create own neural network architecture
- Use information other than image available in vehicle

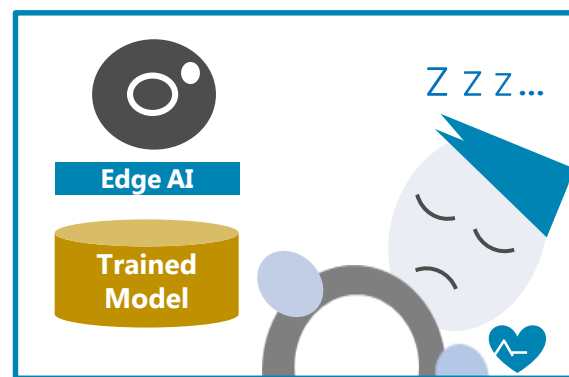
### 2. Performance improvement

- More complicated neural network architecture causes performance issues
- Utilize GPU to accelerate calculations

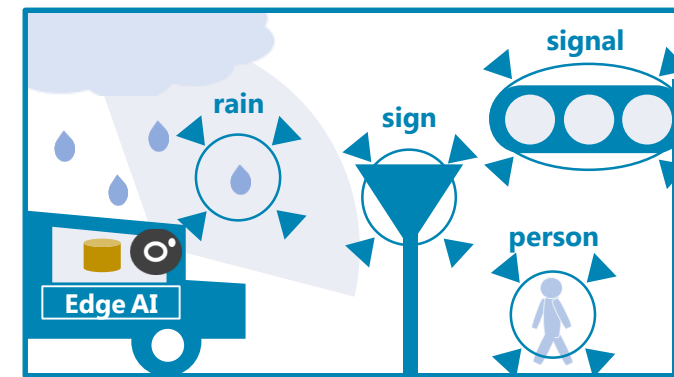
## Future use-cases



Use various information of embedded devices

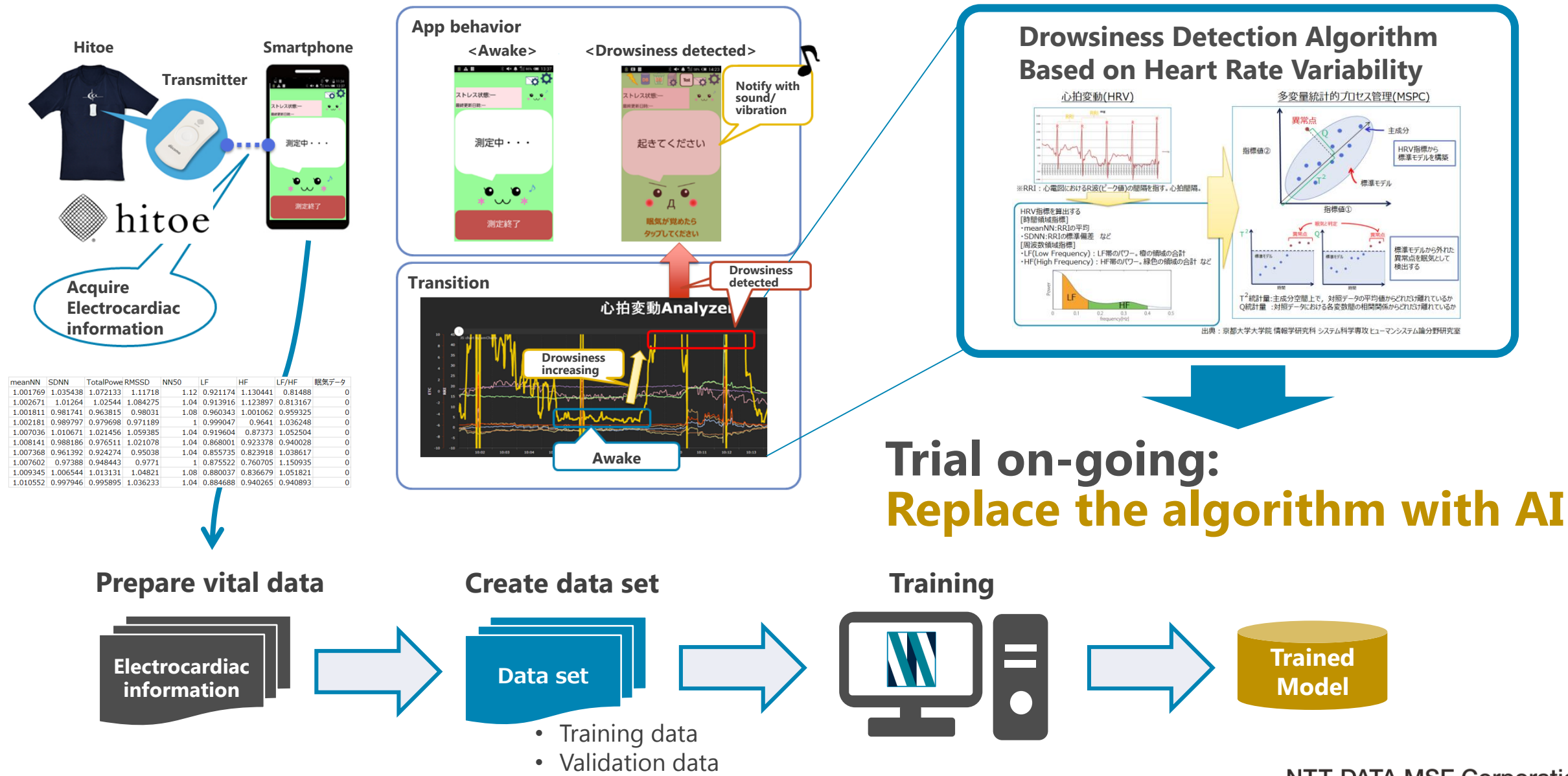


Detect driver's drowsiness



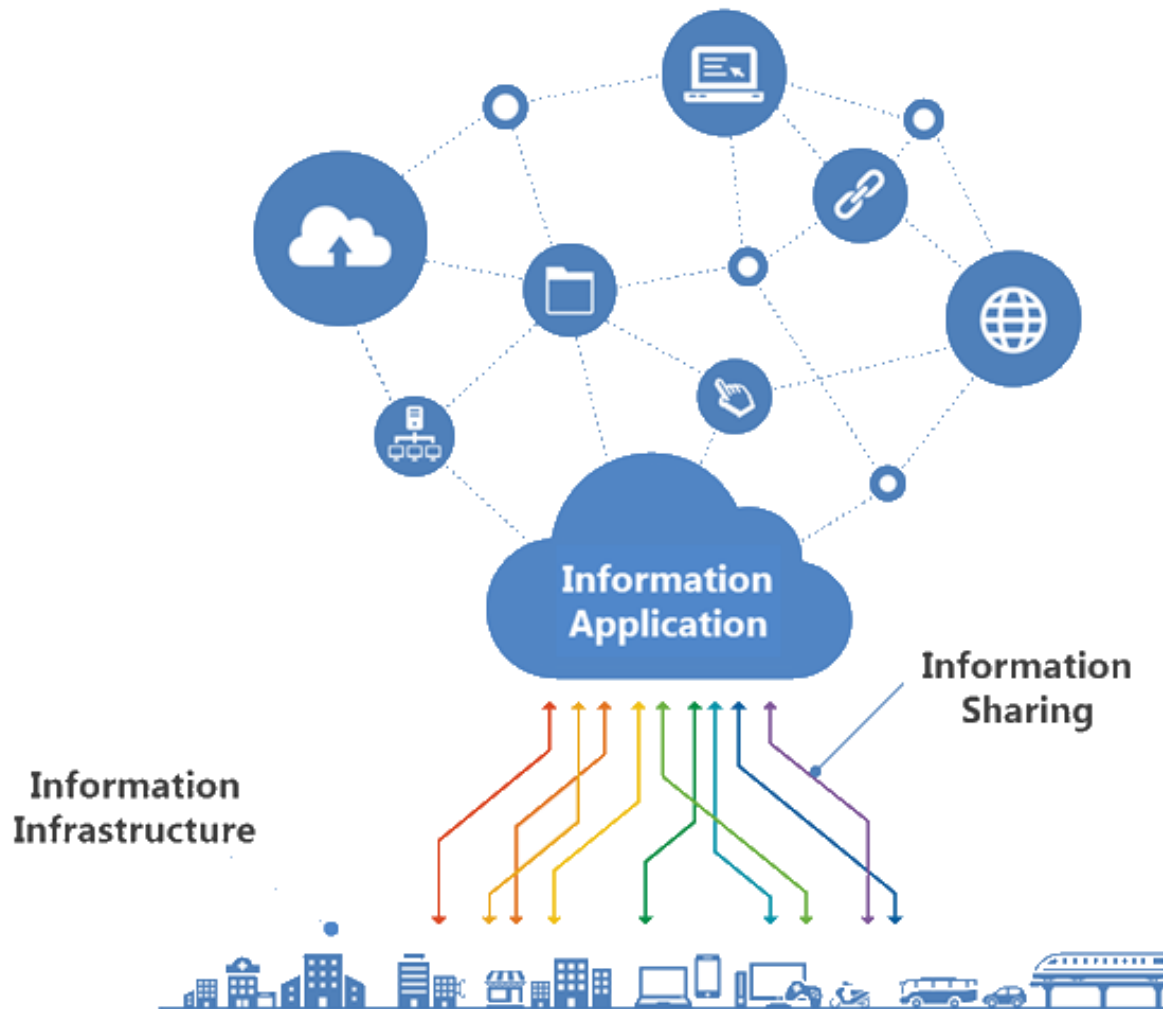
Improvement of autonomous driving technology

# Drowsiness Detection



# Questions?

# Thank you very much!!



## Smart Life Community<sup>®</sup>

